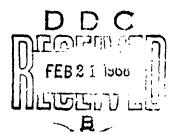
# Technical Memorandum

A COMPUTER PROGRAM FOR THE KOLMOGOROV

GOODNESS OF FIT TEST FOR NORMALITY

Carl B. Bates Jacqueline R. Orsulak

Computation and Analysis Laboratory



U. S. Naval Weapons Laboratory Dahlgren, Virginia

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### U. S. NAVAL WEAPONS LABORATORY

### TECHNICAL MEMORANDUM

January 1968

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# A COMPUTER PROGRAM FOR THE KOLMOGOROV GOODNESS OF FIT TEST FOR NORMALITY

Carl B. Bates and Jacqueline R. Orsulak

Computation and Analysis Laboratory

Approved by:

H. S. OVERMAN, Acting Director, Computation and

Hammon

Analysis Laboratory

While the contents of this memorandum are considered to be correct, they are subject to modification upon further study.

Distribution of this document is unlimited.

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### ABSTRACT

Following a brief discussion on the background, applicability, and limitations of the Kolmogorev Test, a computer program of the Kolmogorev Test, a computer program of the Kolmogorev Test for normality is described. The program is applicable for testing the null hypothesis that a random sample is from a parent normal population whose mean and variance are equal to those of the sample distribution. The program determines the minimum and maximum sample values, computes sample estimates of the mean and variance of the hypothesized normal distribution, and computes the Kolmogorov statistic. Five optional significance levels (0.20, 0.15, 0.10, 0.05, and 0.01) are available, and sample size limitations are  $4 \le n \le 2500$ . An optional feature provides for CRT plot output applicable for both hypothesis testing and interval estimation.

The program is coded in FORTRAN IV for the IBM 7030 (STRETCH) computer.

### FOREWORD

The formulation of the computer program for the Kolmogorov

Test was performed in the Mathematical Statistics Branch of the

Operations Research Division. The coding of the program was performed
in the Operations Sciences Branch of the Computer Programming Division.

Dr. H. W. Lilliefors of the George Washington University supplied the
critical values which are tabulated within the program. The project
was supported under Foundational Research Project No. 29Y, "Computer
Programs for Statistical Analyses."

The date of completion was 15 December 1967.

### 1. INTRODUCTION

The applied statistician is often confronted with the problem of investigating the distribution of sampled data. Because of the desirable characteristics of the normal distribution and because of the frequency of occurrence of these characteristics in physical phenomena, a logical approach to the problem is an examination of the sample distribution to determine if it can be approximated by a normal distribution. This examination is often accomplished by hypothesis testing, i.e., testing the null hypothesis that the randomly observed sample data is from a parent normal population.

The Chi-Square Test originally proposed by Pearson in 1900 appears to be the most popular of the Goodness of Fit Tests. A common criticism, however, of the Chi-Square Test is the loss of power because of the required grouping of the sample data, e.g., see Massey (1951), Birnbaum (1952), or Cochran (1954). In addition, to employ the Chi-Square Test to test for normality, a minimum of four intervals (groups) is needed when the population is not completely specified under the null hypothesis and the two population parameters ( $\mu$  and  $\sigma^2$ ) are estimated by sample statistics. Consequently, the Chi-Square Test is not applicable to very small samples.

An alternative distribution-free test proposed by Kolmogorov in 1933 does not have the objectionable requirements mentioned above for the Chi-Square Test. That is, the Kolmogorov Test requires no grouping of the data and the test is applicable to very small samples. Because the Kolmogorov Test treats individual observations rather than grouped data, sample information may be better utilized by the Kolmogorov Test than by the Chi-Square Test, see Birnbaum (1952). Studies by Massey (1951) and

Kac, Kiefer, and Wolfowitz (1955) indicate that the asymptotic power of the Kolmogorov Test is greater than that of the Chi-Square Test. A recent paper by Slakter (1967), however, shows that the Kolmogorov Test is not uniformly more powerful than the Chi-Square Test. Further studies are necessary, in the opinion of the author of this report, to justify a preference of either test on the basis of power. On the other hand, the definite advantage of the Kolmogorov Test with respect to small samples is sufficient justification for the applied statistician to give serious consideration to the Kolmogorov Test.

### II. THE KOLMOGOROV TEST

### A. Background

The Kolmogorov Test for normality, like the Chi-Square Test for normality, is a test for assessing the agreement of a sample distribution with a parent normal distribution. While the Chi-Square Test is concerned with the agreement of a sample distribution with a theoretical distribution, the Kolmogorov Test is concerned with the agreement of a sample cumulative distribution with a theoretical cumulative distribution.

Consider a random variable X with a continuous cumulative distribution function (cdf), F(x), where

$$P(x) = P\{X \le x\} = \int_{-\infty}^{x} f(x) dx.$$
 (1)

Let  $x_1, x_2, \cdots, x_l$ ,  $\cdots$ ,  $x_l$  be a random sample of n observations (or measurements) of X. The empirical cdf corresponding to the random sample of size n is  $S_n(x)$ , where

$$S_n(x) = \frac{\text{number of } x_1 - x}{n}.$$
 (2)

The Kolmogorov statistic, D(n), is based on the maximum absolute deviation between F(x) and  $S_n(x)$ . That is,

$$D(n) = Max \{F(x) - S_n(x)\}.$$
 (3)

D(n) then is the test statistic to test the null hypothesis of no difference between  $S_n(x)$  and F(x).

Until recently the Kolmogorov Test was applicable for only completely specified F(x) under the null hypothesis. That is, the standard tables, such as Massey (1951), Birnbaum (1952), and Miller (1956), of critical values for testing D(n) are not applicable if population parameters of the theoretical distribution are estimated by sample statistics. Lilliefors (1967) provided a tabulation of critical values applicable when the theoretical distribution is normal and sample statistics are used to estimate the two population parameters. The critical values, which were determined by Monte Carlo calculations, are for five significance levels (0.20, 0.15, 0.10, 0.05, and 0.01) for n ≥ 4. Consequently, the Kolmogorov Test can now be used to examine a sample distribution to determine if it can be approximated by a normal distribution whose mean and variance are equal to those of the sample distribution.

### B. The Kolmogorov Test for Normality

To distinguish the incompletely specified from the completely specified theoretical distribution under the null hypothesis, a circumflex is used denoting the use of sample statistics for the two population parameters of the normal distribution. Equation (1) then becomes

$$\hat{\mathbf{F}}(\mathbf{x}) = \mathbf{P}[\mathbf{X} \leq \mathbf{x}] = \int_{-\infty}^{\mathbf{X}} \hat{\mathbf{f}}(\mathbf{x}) d\mathbf{x}, \qquad (4)$$

where

$$\hat{f}(x) = (1/\sqrt{2\pi}s) \exp[-(x-\overline{x})^2/2s^2],$$
 (5)

and where the familiar statistics, T and s2, are

$$\begin{array}{ccc}
n & n \\
\overline{x} = \sum_{i=1}^{n} / n & \text{and} & s^2 = \sum_{i=1}^{n} (x_i - \overline{x})^2 / (n-1).
\end{array}$$
(6)

Substituting  $\hat{F}(x)$  for F(x) in equation (3), the Kolmogorov statistic becomes

$$\hat{D}(n) = \text{Max} |\hat{F}(x) - S_n(x)|. \tag{7}$$

The null hypothesis that a random sample of n observations is from a parent normal population is tested by comparing  $\hat{D}(n)$  with the critical value,  $D_{\alpha}(n)$ , of TABLE I. If  $\hat{D}(n) > D_{\alpha}(n)$ , reject the null hypothesis at the  $\alpha$ -level of significance; otherwise, do not reject the null hypothesis.

 $\begin{array}{c} \underline{TABLE\ I} \\ \\ D_{\alpha}\left(n\right)\ CRITICAL\ VALUES* \end{array}$ 

Sample	α - Level of Significance				
Size	0.20	0.15	0.10	0.05_	0.01
<u>n</u>	0.20	0.13	0.10		0.01
4	.300	.319	. 352	.381	.417
5	. 285	.299	.315	.337	.405
6	.265	.277	.294	.319	.364
6 7 8 9	.247	.258	.276	. 300	.348
8	.233	.244	.261	.285	.331
9	.223	.233	.249	.271	.311
10	.215	.224	. 239	. 258	.294
11	.206	.217	.230	. 249	.284
12	.199	.212	.223	. 242	.275
13	.190	. 202	.214	. 234	.268
14	. 183	.194	.207	.227	.261
15	.177	.187	.201	.220	.257
16	.173	.182	. 195	,213	.250
17	.169	.177	.189	.206	.245
18	.166	.173	.184	.200	.239
19	.163	.169	.179	.195	.235
20	.160	.166	.174	.190	.251
25	. 1.42	.147	. 158	.173	200
30	. 1.31	.136	.144	.161	. 187
Over 30	. 736	.768	.805	. 886	1.031
	$\sqrt{n}$	√n	√n	, n	√n

\*From Lilliefors (1967), On The Kolmogorov-Smirnov Test for Normality with Mean and Variance Unknown, Journal of the American Statistical Association, Vol. 62, pp. 399-402 and personal communication of October 16, 1967

The Kolmogorov Test is a two-sided test in that departures of  $S_n(x)$  in either direction from  $\hat{F}(x)$  increase  $\hat{D}(n)$ . The rejection region of  $\hat{D}(n)$  is the region outside of the band  $\hat{F}(x) + D_0(n)$ , i.e.,

$$F(U) = \hat{F}(x) + D_{\alpha}(n)$$

$$F(L) = \hat{F}(x) - D_{\alpha}(n).$$
(8)

That is, the null hypothesis is rejected at the  $\alpha$ -level of significance if  $S_n(x)$  passes outside this band. Because  $S_n(x)$  is a step-function,  $S_n(x)$  is constant for  $\mathbf{x}_{(j-1)} \leq \mathbf{x} < \mathbf{x}_{(j)}$ , where  $\mathbf{x}_{(j)}$  is the j<sup>th</sup> order statistic. Therefore, if  $S_n(x)$  passes into the lower rejection region,  $\hat{D}(n) = |\hat{\mathbf{F}}(\mathbf{x}_{(j)}) - S_n(\mathbf{x}_{(j-1)})| \geq D_q(n)$ , see Figure 1(a); if  $S_n(x)$  passes into the upper rejection region,  $\hat{D}(n) = |\hat{\mathbf{F}}(\mathbf{x}_{(j)}) - S_n(\mathbf{x}_{(j)})| \geq D_q(n)$ , see Figure 1(b).

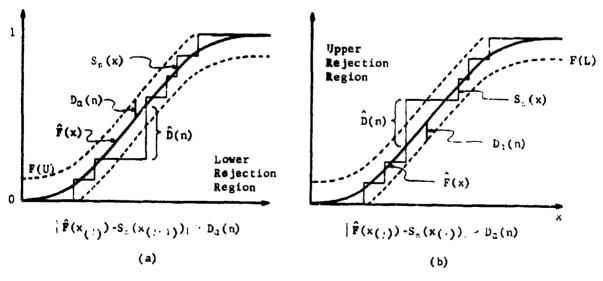


Figure :

In addition to the utility of  $D_{\alpha}(n)$  for hypothesis testing,  $D_{\alpha}(n)$  may also be used for interval estimation. The  $100(1-\alpha)\%$  confidence limits for the "true" cumulative distribution function, F(x), are

$$S(L) < F(x) < S(U), \tag{9}$$

where

$$S(U) = S_n(x) + D_Q(n)$$
  
 $S(L) = S_n(x) - D_Q(n)$ . (10)

### III. COMPUTATIONAL PROCEDURE

Before performing the computations described in the pravious section, the n sample values are transformed by one of the following thirteen available transformations.

Transformation Number	Transformation
1	x <b>←</b> x
2	$x \leftarrow \ln x$
3	$x \leftarrow \ln(\ln(x))$
4	$x \leftarrow \ln(A+x)$
5	$x \leftarrow ln(B+ln(C+x))$
6	<b>x</b> ← √ <b>x</b>
7	$x \leftarrow 1/x$
8	$x \leftarrow 1/(D+x)$
9	x + sin x
10	$x \leftarrow 2 \sin^{-1}/x$
11	x ← x/E
12	x + sin x
13	x + cos x

The transformations are identified on card type 3, and the constants A,B,C,D, and E are input on card type 5 (see Section IV).

After ordering the n transformed sample values, the subset

$$x_{(1)} < x_{(8)} < \cdots < x_{(1)} < \cdots < x_{(k)} ; k \le n$$
 (11)

of sample data is considered. The empirical cdf,  $S_n(x)$ , is evaluated for each of the k unique sample values. To evaluate  $\hat{F}(x)$ , the k unique sample values are standardized by substituting  $u = (x-\overline{x})/s$  in  $\hat{f}(x)$ . The standardized pdf,

$$\hat{\phi}(u) = (1/21) \exp[-\frac{1}{2}u^2],$$
 (12)

is then used to evaluate  $\hat{f}(x)$  which becomes

$$\hat{\mathbf{F}}(\mathbf{x}) = \hat{\boldsymbol{\Phi}}(\mathbf{u}) = \int_{-\infty}^{\mathbf{u}} \hat{\boldsymbol{\phi}}(t) dt.$$
 (13)

In addition to  $x_{(1)}$ ,  $x_{(k)}$ ,  $x_{(k)}$ ,  $x_{(1)}$ , n,  $\overline{x}$ ,  $s^2$ , and s, system output consists of u,  $\hat{\mathbf{F}}(\mathbf{x})$ ,  $\mathbf{S}_{\mathbf{h}}(\mathbf{x})$ , and  $|\hat{\mathbf{F}}(\mathbf{x}) - \mathbf{S}_{\mathbf{h}}(\mathbf{x})|$  for each of the k unique order statistics along with the identification of  $\hat{\mathbf{D}}(\mathbf{n})$ . CRT output consists of one or more of the four following plot types.

Plot Type A - 
$$S_n(x)$$
,  $\hat{F}(x)$ ,  $F(U)$ ,  $F(L)$ 

Plot Type B -  $S_n(x)$ ,  $\hat{F}(x)$ ,  $S(U)$ ,  $S(L)$ 

Plot Type C -  $S_n(x)$ ,  $S(U)$ ,  $S(L)$ 

Plot Type D -  $S_n(x)$ ,  $\hat{F}(x)$ 

The desired plot type(s) are identified on card type 3.

### IV. INPUT PREPARATION

### A. Deck Setup

The input deck is listed below by card type. Multiple jobs may be processed by stacking card types 1 through 6.

CARD TYPE 1 - JOB IDENTIFICATION CARD

CARD TYPE 2 - VARIABLE FORMAT CARD

CARD TYPE 3 - MAIN CONTROL CARD

CARD TYPE 4 - ALPHA IDENTIFICATION CARD

CARD TYPE 5 - TRANSFORMATION CONSTANT CARD

CARD TYPE 6 - SANDLE DATA CARD

### B. Input Deck Description

12

1-2

Entries for variables with an I format specification must be right-adjusted in the specified field.

### CARD TYPE 1 - JOB IDENTIFICATION CARD

Column	Format	<u>Variable</u>	Description
1-80	1QA8	<b>J08(1)</b> - J08(10)	Job Identification
	CARD TYPE 2	- VARIABLE FORMAT CARD	
Column	Format	Variable	Description
1-80	10A8	FMT(1) - FMT(10)	Format for reading in the sample data. The format specifications must be enclosed in parentheses.
	CARD TYPE 3	- MAIN CONTROL CARD	
Column	Format	<u>Variable</u>	Description

NRUN

Number of transformations

to be performed (1 \le NRUN \le 13)

Column	Format	<u>Variable</u>	Description
11	11	itran(1)	<pre>0 - Do not perform transformation number 1 1 - Do perform</pre>
			transformation number 1
12	Iì	itran(2)	O - Do not perform transformation number 2 1 - Do perform transformation number 2
	:	:	:
:	:	:	:
:	:	:	:
23	11	itran (13)	0 - Do not perform transformation number 13 1 - Do perform transformation number 13
25	12	MEDIUM	<ul> <li>2 - Sample data is</li> <li>input on cards</li> <li>1 - Sample data is</li> <li>input on tape</li> </ul>
31	12	IA	<ul><li>0 - Do not perform</li><li>plot type A</li><li>1 - Do perform</li><li>plot type A</li></ul>
32	12	138	<ul><li>0 - Do not perform</li><li>plot type B</li><li>1 - Do perform</li><li>plot type B</li></ul>
33	12	IC	O - Do not perform plot type C l - Do perform plot type C
34	12	10	O - Do not perform  plot type D  1 - Do perform  plot type D
40-41	12	₩øtic	Number of desired tick marks on the abscissa axis (if left blank NØTIC is set equal to 15)

Column	Forma:	Variable	Description
44-45	12	IRA(1)	Number of decimal digits to be used in labeling abscissa tick marks for transformation number 1
46-47	12	IRA(2)	Number of decimal digits to be used in labeling abscissa tick marks for transformation number 2
:	:	:	:
<b>:</b>	:	:	;
:	:	:	ı
68-69	12	IRA (13)	Number of decimal digits to be used in labeling abscissa tick marks for transformation number 13
74-75	12	NBR	Number of sample data per card
78	rı	NC	O - Card type 5 is not input 1 - Card type 5 is input

# CARD TYPE 4 - ALPHA IDENTIFICATION CARD

To obtain plots, at least one alpha must be indicated.

		- Francisco.	
Column	<u>Pormat</u>	<u>Variable</u>	Description
1	11	IALPHA (1,1)	<pre>1 - Alpha = .20 is desired for transformation number 1 0 - Alpha = .20 is not desired for transformation number 1</pre>
2	Il	IALPHA(2,1)	<pre>1 - Alpha = .15 is desired for transformation number 1 0 - Alpha = .15 is not desired for transformation number 1</pre>

Column	Format	<u>Variable</u>	Description
3	T1	IALPHA(3,1)	<pre>l - Alpha = .10 is desired for transformation number l 0 - Alpha = .10 is not desired for transformation number l</pre>
4	11	IALPHA (4,1)	<pre>l - Alpha = .05 is desired for transformation number 1 0 - Alpha = .05 is not desired for transformation number 1</pre>
5	11	IALPHA (5,1)	1 - Alpha = .01 is desired for transformation number 1 0 - Alpha = .01 is not desired for transformation number 1
6	11	IALPHA(1,2)	1 - Alpha = .20 is desired for transformation number 2 0 - Alpha = .20 is not desired for transformation number 2
:	:	:	:
:	:	:	:
65	11	IALPHA(5,13)	<pre>1 - Alpha = .01 is des.red for transformation number 13 0 - Alpha = .01 is not desired for transformation number 13</pre>

### CARD TYPE 5 - TRANSFORMATION CONSTANT CARD

This card type is omitted if NC = 0 on card type 3.

Column	Format	Variable	Description
1-14	F14.6	A	Constant for transformation number 4
15-28	F14.6	В	Constant for transformation number 5
29-42	F14.6	C	Constant for transformation number 5
43-56	F14.6	D	Constant for transformation number 8
57-70	F14.6	I .	Constant for transformation number 11

### CARD TYPE 6 - SAMPLE DATA CARD(S)

The sample size (n) must be  $4 \le n \le 2500$ . This card type is omitted if MEDIUM = 1 on card type 3.

Column	Format	<u>Variable</u>	Description
1-2	12	ITEST	Blank - This is not the last card containing sample data \$\frac{2}{40}\$ - Number of sample data on the last card or record
3-80	Variable	xx	The array of sample data which must be input according to the variable format on card type 2

### C. Request Sheets

If the sample data (card type 6) is input on purched cards, the job request sheet is prepared as shown below.

	ğΗ V.	J.	025	MAK				A3		A141		7309	SF TUP &	01
COMPILE 2 GO	<b>□</b> 551				DE	CARD	MT.	> * 0 6	MANNER	300 Y I	Cast			
C0M+1L 60	r kop	2	1	5	8	K	8	Δ	3	FST, COMP	ILEP TIME	30 \$	litinki tibe EC.	12/5/6
			TAPE	S CAL	LED	FOR	BY PF	OBL	M PR	MARRO				·····
TAPE NUMBER	Scrate	h			·····									
FILE PROTECT ON						<u> </u>							1	
PROGRAMMER NUMBER														
SPECIAL MANULING														
ABEOJ HOLD	D,	.oup				D. BEG ADDITI	MEVE	RSE COMME	NTS.			· · · · · · · · · · · · · · · · · · ·		

7030 JOB REQUEST NOW-NWL-5230/29 (REV. 11-66)

If the sample data is input on tape, the tape number is written on the job request sheet instead of "Scratch". And the "REEL, NUL" card (second card) of the IOD deck is prepared as follows.

### Column

1 Punch the letter "B".

10-17 Punch "REEL, NUL" (or "REEL, PUL" if the tape is file protected).

18-21 Punch the number of the tape. The CRT request sheet is prepared as follows.

TRAID CAMERA OUTPUT FILM OR PAPER COPIES NOW-HWL-SROD 2 (REV. 8-84)	REQUEST				
PROGRAMMER	ROOM	BLDG.	PHONE	DATE	
J. ORSUŁAK	A141	1200	7309	12/5/67	
FILM IDENTIFICATION		<del></del>			
K5 -A3					<b>.</b> .
APPROXIMATE NUMBER OF FRAMES					
The total number of plots requested	for the ic	bs being	process	ed.	
NUMBER OF PAPER COPIES PER FRAME					
The number of copies desired of eac	h plot.				
FOR C	PERATORS OF	NLY			
COUNTER READING START	FINISH				
DATE AND TIME PROBLEM MAN					

V. OUTPUT FORMAT

A. System Output

(Job identification as given on card type 1)

FORMAT FOR SAMPLE DATA

KO, OF RUNS = \_\_\_ IRUN = \_\_\_

TRANSPORMATION TRANSFORMATION 3 = TRANSFORMATION TRANSFORMATION 1

TRANSFORMATION TRANSPORMATION 7 TRANSFORMATION 6 = TRANSFORMATION TRANSPORMATION 12 = TRANSPORMATION 11 = TRANSFORMATION 10 = TRANSFORMATION 9 =

TRANSFORMATION 13 = MEDIUM =

PLOT TYPE B = PLOT TYPE C = PLOT TYPE D =

NBR = \_\_\_\_\_ NC - \_\_\_\_\_ ITEST = \_\_\_

A:PHA IDENTIFICATION =

TRANSFORMATION CONSTANT CARD A = B =

ا ن

16

PLOT TYPE A =

TRANSFORMED DATA	`X	 , x,	 , x
ORIGINAL DATA	×S	 ×.	 x <sup>u</sup>
TRANSPORMED DATA	ĸ1,	 K'-1	 $x_{r-1}'$
ORIGINAL DATA	×	 K <sub>1</sub> - 1	 $\mathbf{x}_{n-1}$

RANGE =  $\frac{x'(z)^{-x'}(1)}{|x'(z)|}$ 

HAX IMUM ==

MEAN =

\*INIMIM =

SAMPLE SLES =

# KOLMOGOROV TEST FOR NORMALITY

VALUE		•			(1-1)					بي ا
ABSOLUTE VALUE P(J)-S(J)	(F(1)-S(1)	••	••	••	F(1-1) =	(1)-S(1)	••	••	**	P(4) -S(4)
ABSOLUTE VALUE F(J)-S(J-1)	F(1) -S(0)	14	••	••	F(1-1) -8(1-2)	F(1) "S(1-1)	••	.,	••	P(x) -S(x-1)
OBSERVED CDF S(J)	S(1)	••	••	••	8(1-1)	8(1)	••	••	••	S(k)=1
THEORETICAL CDF F(J)	P(1)	••	••	••	P(1-1)	F(1)	••			<b>P</b> (k)
STANDARDIZED Variate U(J)	(t) <sub>n</sub>	••	••	••	(t-i) <sub>h</sub>	(c) <sub>a</sub>	10	••	••	(k) <sub>n</sub>
FREQUENCY	<b>f</b> (ι)	••	••	••	f(1-1)	(1)	••	••	••	f(x)
TRANSPORMED DATA X(J)	κ(ι) <sub>×</sub>	••	••	••	x(1-1)	x(1)	••	••	••	x(x)
ORDER INDEX (J)	end.		••	••	j-1	7	••	••		포

 $D(ORDER INDEX) = \hat{D}(n)$ 

Note:  $\hat{f}(z) = \hat{f}(x'(z))$ ,  $\hat{F}(z) = \hat{F}(x'(z))$ , and  $S(z) = S_n(x'(z))$ 

### B. CRT Output

The plots are performed by the system "GRF Plot Subroutine" and are plotted on the CRT printer units. The plots are in an ordinary Cartesian coordinate system with origin  $(x'_{(1)},0)$ . The abscissa of the system is the x'-axis. A zero (0) is plotted at the point  $(\overline{x}',0.5)$ . The four types of plots are labeled as follows.

Plot Type A -  $S_n(x)$ ,  $\hat{\mathbf{f}}(x)$ ,  $\mathbf{F}(U)$ ,  $\mathbf{F}(L)$ 

OBSERVED RELATIVE CUMULATIVE FREQUENCY, THEORETICAL RELATIVE CUMULATIVE FREQUENCY, AND CRITICAL REGION BOUNDS FOR ALPHA =  $\alpha$  LEVEL OF SIGNIFICANCE, D(ALPHA) =  $D_{\alpha}(n)$ 

Plot Type B -  $S_n(x)$ ,  $\hat{F}(x)$ , S(U), S(L)

OBSERVED RELATIVE CUMULATIVE PREQUENCY, THEORETICAL RELATIVE CUMULATIVE PREQUENCY,  $\frac{100(1-\alpha)}{\text{CUMULATIVE}} \text{ PERCENT CONFIDENCE BANDS FOR THE THEORETICAL CUMULATIVE DISTRIBUTION FUNCTION,} \\ D(ALPHA) = D_{\alpha}(n)$ 

Plot Type  $C - S_n(x)$ , S(U), S(L)

OBSERVED RELATIVE CUMULATIVE FREQUENCY AND THE  $100(1-\alpha)$  PERCENT CONFIDENCE BANDS FOR THE THEORETICAL CUMULATIVE DISTRIBUTION FUNCTION

Plot Type D -  $S_{p}(x)$ ,  $\hat{F}(x)$ 

OBSERVED AND THEORETICAL RELATIVE CUMULATIVE PREQUENCY

### C. Program Running Time

The running time is printed at the end of each job processed. The following table of observed times may be used as a guide for estimating program running times.

Sample Size	No. of Runs	Total No. of Plots	Time in Seconds
25	1	4	12 .4
140	1	4	16.1
10	1	6	19.0
30	1	6	20.0
20	2	8	25 .4
20	4	16	45 .8
20	5	20	56 .4
1000	1	2	54.6

### VI. EXAMPLE PROBLEM

### A. Problem Description

The example problem consists of 20 observations ( $x_1$ -values) randomly sampled from a continuous distribution. Two runs were performed, one on the original observations and one on the logarithmically transformed values. Because the minimum original sample value was -750, transformation number 4 was used with the constant A = 751. All four available plot types were performed. The 0.05-level of significance was chosen for testing the null hypothesis of normality.

The input deck setup is illustrated on the following Data Card Layout Sheet.

B. Input for Example Problem

					ATAC	A CABO LAVORT	N. Carr								PROG	3
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1	<del> </del>															
1		130		-130		-	24	Ė	-194.6		246-		667-	] [6]		<u> </u>
í		-130		1745	-	-	50	E			3.7.8.		6.4			
1_		7.53	-	12.5		1-7	59.	Ë	-490		-624				-	
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l										-						

# C. Program Output

EXAMPLE PROBLEM
FORMAT FOR SAMPLE DATA (12.7F.10.5)

NO. OF RUSE = 2

IRANSFORMATION 1 = 1 TRANSFORMATION 2 = 0 TRANSFORMATION 3 = 0 TRANSFORMATION 4 = 1

TRANSFORMATION 1 = 0 TRANSFORMATION 0 = 0 TRANSFORMATION 1 TRANSFORMATION 1 = 0 TRANS

130.000000000 - 424.0000000000 - 742.0000000000 - 738.000000000 - 738.000000000 - 775.000000000 - 775.000000000 - 778.000000000 - 778.000000000 - 778.000000000	15250000€+04
28.4.4.4.4.4.4.4.4.4.4.4.4.4.4.4.4.4.4.4	RANGE = .152
0R161NAL DATA 130-000000000 -424.0000000000 -742.0000000000 -718.000000000 -775.000000000 -733.000000000 -734.0000000000	
18ANSFORMED DATA -143.0000000000 -130.00000000000000000000000000000000000	= MEAN =
0R 16 1NAL BATA -743.0000000000 -743.0000000000 -743.0000000000 -693.0000000000 -750.0000000000 -750.0000000000 -750.0000000000 -750.0000000000 -746.00000000000 -746.00000000000	HINIMAN =7500000E+GS SAMPLE SIZE = 20 VARIANCE = 16005451E+TE

KOLNOGGROV TEST FOR NORMALITY

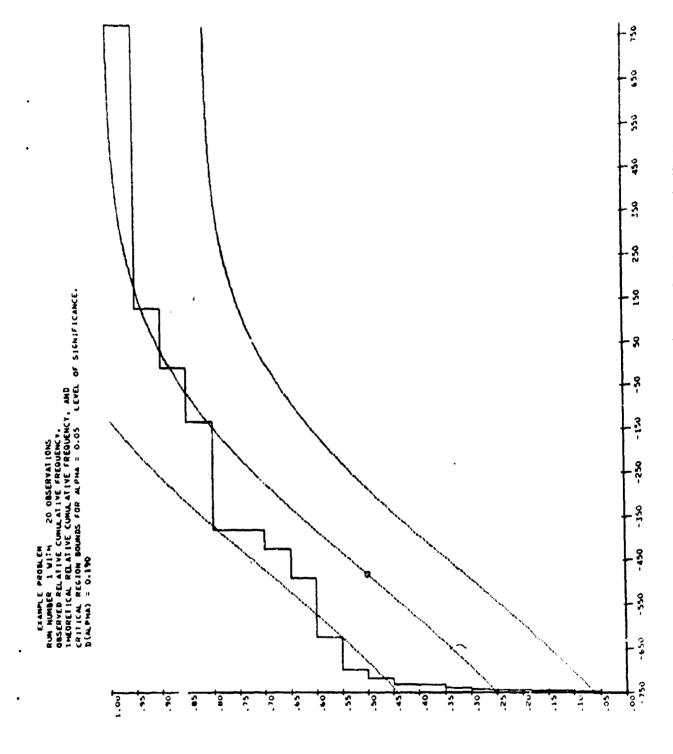
GRDER INDEX	TRANSFORMED DATA X(J)	FREGUENCY	STANDARDIZED THERRETICAL CDF VARIATE U(J) F (J)	F OBSERVED COF S(J)	ABSOLUTE VALUE F(J)-S(J-1)	AE50LUTE VALUE F (1)~5 (1)
-	-750.00000		67133468619 0.251005376963	3 0.050000000000 .25i0053769533	.25:0053769633	.2010053769633
۰ ~	-746.000000	_	1.06136126 0.05041414190.		5006156166151	1044419519005
pa,	-745.009900	- 0	65388123103 0.2%595807833	. •	1065958078329	.0065958078329
❤ 1	- 745 (00000)	<b>4</b>	-,65138788029 0.257399701526	٠.	7.30000000000 .0073997015265	09/13/07/889647
າພ	-739.000000	<b>.</b>	54390782808 G-259819211033	•	0.550000000000	1353068062717
	-733.000000		0.626947/2360 0.444633193/-	_	. 4500000000000 . 128524952909	.1828524952903
Œ	-730.000000			_	129217577448	.2229217577446
<b>o</b>	-718.000000	-1 -	54417779856 0.293162163623	_	0.53000000000 .2056378363766	2568378363765
01	-699.00000	-	35717249321 0.360481896148		1695161038518	*140101814455.
= 2	-490.00000			0	5.6500000000000000000000000000000000000	1437386369
2 12	-424.000000		G.14149765438 0.556261595059	<i>-</i>	.0969017176364	1989017176.04
14	-378.000000	N ·	4.236191.0002.00010102.00 410102.000000.00000000000000000000000000	-		.040913576065
15	-130.00000	<b></b> 4 .	0.8/404/// 54 0.800/404/02/914/4	_		.0192547720667
16	-8.000000		CHARLES O OCENTRATES.			0139045099075
1.7	130.00000	P	1.326133636 0.350350704 1.11102518918 0.959126562704		1.000000000000 .0491289627045	.0006710372955
ž		•				

RANGE OVERFLON \*
RANGE OVERFLON
HANGE OVERFLON
RANGE OVERFLON
RANGE OVERFLON
RANGE OVERFLON

\*RANGE OVERFLOW indicates one or more points are not within the ranges of the axes.

HANSE OVERFLOW

D( 10) = 0.2568

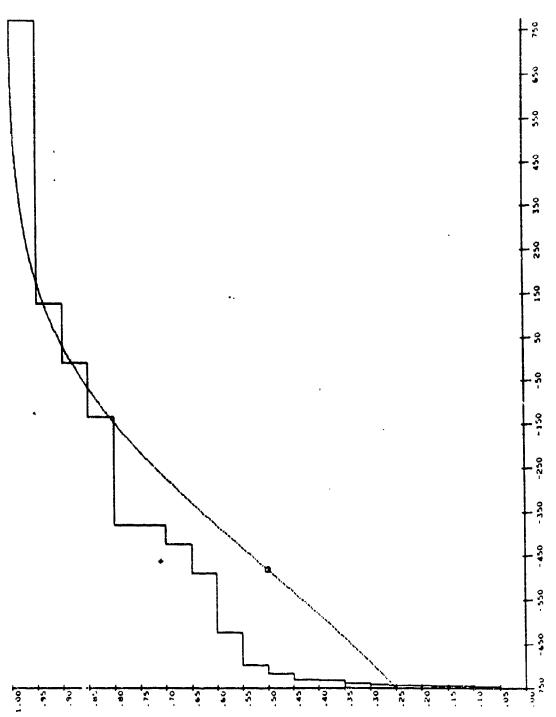


. 20 ENAMPLE PROBLEM
RUM MANBER 1 MITM 20 OBSERVATIONS
OBSERVED REALIVE COULLATIVE FREQUENCY.
1 HEGRETICAL RELATIVE CHULATIVE FREQUENCY.
1 PERCENT COMFIDENCE DANDS FOR THE THEORETICAL CUMLATIVE DISTRIBUTION FUNCTION.
D (ALPMA) = 0.190 220 **,** 2 Ş , K.O. 0. . 35 05. 4 . 1.001 .63 .50 . 15 6 96. 5 30 . 75 2 3 8

13

1 % . . 3 EXAMPLE PROBLEM
RUM NUMBER I WITH ZO OBSERVATIONS
OBSERVED RELATIVE CUMULALIVE FREQUENCY AND THE
95 PERCENT CONFIDENCE BANDS FOR THE THEORETICAL CUMULATIVE DISTRIBUTION FUNCTION 350 . **9** - 8 - Ş - Ş 3. -350 450 0 - 550 -630 β<mark>eε.</mark> 104. 3 ć Č. 0 ... ٠,٥ 59. .45 .30 -20 . 10 3. .13 \$0.

ERANPLE PROBLEM PO DESCRYATIONS OBSERVED AND THEORETICAL RELATIVE COMULATIVE FREGUENCY 1.30 56.



TRANSFORMATION 4 = 1	TRANSFORMATION 8 = 0 TRANSFORMATION 12 = 0	PLOT TYPE D = 1	00000 00000 00000 00000
	TRANSFORMATION 7 = 0 TRANSFORMATION 11 = 0	MEDIUM = 2 PLOT TYPE B = 1 PLOT TYPE C = 1 IRA = 0 0 2 0 0 0 0 0 0 0 AF = 1 FIST = 6 .	APPLA IDENTIFICATION = 00010 00000 00000 00010 00000 00000 00000 00000 00000 00000 0000
(12, F 10.5) RUM = 2 TRANSTONATION 2 = 0	TRANSFORMATION 6 = 0 TRANSFORMATION 10 = 0	PEDION = 2 PLOT TYPE B = 1 IRA = 0 0 0 2 0	00010 0000 0000 00010 T CARD A = 751.0000 D = .0000
EXAMPLE PROPLEM FORMAT FOR SAMPLE DATA NO. OF RUNS. 2 INUM. 2 TOANSEEDHATION TRAIN	TRANSFORMATION 5 = 0	TRANSFORMATION 13 = 0 PLOT TYPE A = 1 NOTIC = 15 NOTIC = 15	ALPHA IDENTIFICATION =

TRANSFORMEE DATA 6.7810576259 5.789601709 2.197224573 3.4965075615 1.7917594632 7.3304052118 5.9215784196 2.8903717579 2.4849066498 4.8441870865	.733040%E+01
	RANGE :
08161NAL BATA 130.00000000000000000000000000000000000	.73304012E+01 .40410096E+01 .21420897E+01
TRANSFORMED DATA 2.0794415417 6.4313310819 2.0794415417 3.9512437186 3.0445224377 0.0000000000000000000000000000000000	HAXIMA = NEAN = NEAN = STANDARD DEVIATION =
9R16 INAL DATA -743.0000000000 -130.0000000000 -730.0000000000 -750.0000000000 -750.0000000000 -746.0000000000 -746.0000000000	MINIMUM = .43231173E-14 SAMPLE SIZE = 20 VARIANCE = .45885482E+01

OR MOCCORON TEST FOR NORMALITY

			KOLMOGOROV IEST FOR NORMALITY	New July			
GRDER INDEX (J)	TRANSFORMED DATA X (J)	FREQUENCY	STANDARDIZED THEORETICAL COF	OBSERVED CDF S(J)	ABSOLUTE VALUE F(J)+S(J+1)	ABSOLUTE VALUE F (J) = S (J)	
, pad	0.00000		-1.88647964775 0.029616622275	0.05000000000	.0296166222748	.020383377252	
~ '	1.609436	<b></b> -	-1.13513952940 0.128160693262 -1.05002565268 0.146855459970	-		.0031445400302	
v) #3	2.079442	• ~	91572591940 0.179907588464		.0299075884644	.0100924113336	
· vn	2.197225	<del>, .</del> .	86074080984 0.194692592440	0.300000000000000000000000000000000000		1162139084254	
٦	2.484907	,,,	-,756410/63/ 0.295381045959 -,53715623373 0.295381045959			1044189544409	
- 10	3.044522		-	ō (	.0791026295686	1291026292600	
· σ·	3.496508	<del>,</del> ,	25419151488 0.399674143108	9 0.300000000000000000000000000000000000		.0667127814236	
01	3.951244			,		.0461509097841	
	5.564520	ھني د		•	-	.1:15263459866	
2	5.789960	-4	_	0.700000000000	100025067113	.0100025067313	
71	5.921578	∾ •	0.87791368234 0.6100000000000000000000000000000000000		-	.0177617200130	
15	6.431331	p=4 p=	1.11268346717 0.007781751445 1.19041712570 0.884853751445	ó	•	.0151462485546	
<u>e</u> !	0.010.0		1 2791476-BORT 0.899575335841	-		.0504246641593	
2 2 2	7.330405	٠		1.00000000000000	.0123199126895	.0623199128895	
D( 12) = 0.1615		٠					
RANGE OVERFLOW		٠					
RANGE OVERFLOW							
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25.39

I INC IN SECONDS =

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EXAMPLE PROBLEM

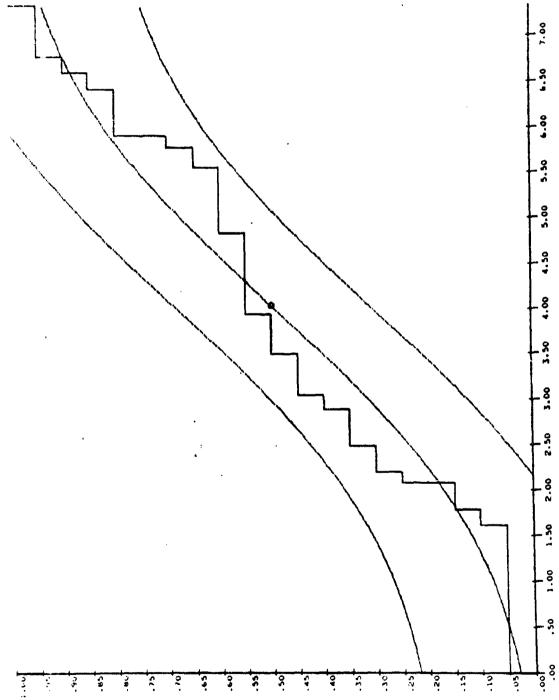
RUN NUMBER 2 WITH 20 OBSERVATIONS

OBSERVED RELATIVE COMULATIVE FREQUENCY.

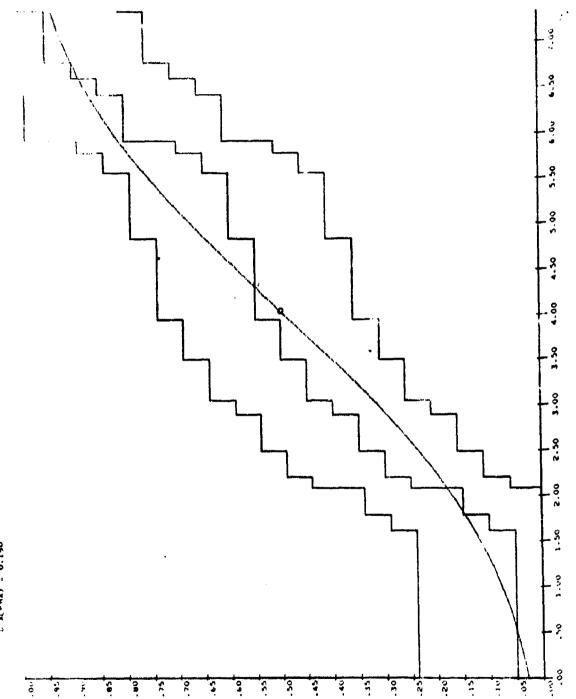
THEORETICAL HELATIVE CUMULATIVE FREQUENCY, AND

CHITCAL REGION BYUND: FOR ALPHA = 0.05 LEVEL OF SIGNIFICANCE.

D(ALPHA) = 0.190



EXAMPLE PROBLEM
RUM HAMBER 2 MITH 20 0BSERVATIONS
GDSERVED RELATIVE CUMILATIVE FREQUENCY.
THEORETICAL RELATIVE CUMINATIVE FREQUENCY.
95 PERCENT CONFIDENCE BANDS FOR THE THEORETICAL CUMILATIVE DISTRIBUTION FUNCTION.
C"ALPHA) = 0.190



7.00 £.00 5.50 EXAMPLE PROBLEM
RUN NUMBER 2 WITH 20 OBSERVATIONS
OBSERVED RELATIVE COMMEATIVE FREQUENCY AND THE
95 PERCENT COMFIDENCE BANDS FOR THE TMEGRETICAL CUMILATIVE DISTRIBUTION FUNCTION 8.8 4.50 4.00 3.50 .00 8. % 5.00 1.50 1.00 ş 100.. 1.00 09. ô. 0 \*<u>\*</u> , v .63 Š \$ 0. .35 .30 .95 Š Š .20 -15 .05

3.8 1.00 . 0 5. .65 .60 3 . 35 30 +--.75 .55 3 Ç 1.5

## D. Discussion of Test Results

To test the null hypothesis "The 20 randomly sampled observations are from a parent normal population" at the 0.05-level of significance,  $\hat{D}(n) = 0.2568$  (from the first run) is compared with  $D_{0.05}(20) = 0.190$ . The null hypothesis is, therefore, rejected at the 0.05-level of significance. Note that the maximum absolute deviation appears at  $x_{(10)} = -699$ , and at this point  $S_{a}(x)$  has passed above F(U) into the upper rejection region (see pages 24 and 25). To test the null hypothesis of normality of the transformed random variable at the 0.05-level of significance,  $\hat{D}(n) = 0.1615$  (from the second run) is compared with  $D_{0.05}(20) = 0.190$  (see page 31); consequently, the null hypothesis is not rejected. Unlike run number 1, plot type A for run number 2 on page 32 shows  $S_{n}(x)$  remaining within the "acceptance" band. Therefore, we conclude at the 0.05-level of significance that the original random variable is approximately log normally distributed.

```
VII.
    PROGRAM LISTING
9
         TYPE.COMPILGO.FORTRAN.PM
T
         SUBTYPE.FIOD
81 D
         IOD.TAPE...EVEN..SAVE
         REEL . NULXXXX
8
         100 . BREADER
         END
         SUBTYPE FORTRAN . LMAP . PUNCH
¢
      KOLMOGOROV TEST FOR NORMALITY
      DIMENSION TABLE (28, 5)
      DIMENSION JOB(10), FMT(10), IFILE(35), ITRAN(13), IALPHA (5,13),
     1 X(2500), U(2500), ALPHA(5), IFREQ(2500), TEN(20), T(2501), IRA(13
     2 ), xx(2500), s(2500), sn1(2500), sn(2500), f(2500), fn(500),
     3 FL(500), FU(500), SL(2500), SU(2500)
      COMMON /PRIME/ XP(2, 1250), XXP(2, 1250)
      EQUIVALENCE (IFILE(1). NRUN). (IFILE(2). ITRAN(1)).
     1 (IFILE(3): ITRAN(2)): (IFILE(4): ITRAN(3)): (IFILE(5): ITRAN(4)):
     2 (IFILE(6): ITRAN(5)): (IFILE(7): ITRAN(6)): (IFILE(8): ITRAN(7)):
     3 (IFILE(9): ITRAN(8)): (IFILE(10): ITRAN(9)): (IFILE(11): ITRAN(10
       )). ([FILE(12). [TRAN(11)). ([FILE(13). [TRAN(12)). ([FILE(14).
     5 ITRAN(13)) + (1FILE(15) + MEDIUM) + (1FILE(16) + 1A) + (1FILE(17) + 18)
                           (IFILE(19), ID), (IFILE(20), NOTIC)
     6 . (IFILE(18), IC).
      EQUIVALENCE (IFILE(21), IRA(1)), (IFILE(22), IRA(2)), (IFILE(23),
     11RA(3)) + (IFILE(24) + IRA(4)) + (IFILE(25) + IRA(5)) + (IFILE(26) +
     21RA(6)). (1F1LE(27). 1RA(7)). (1F1LE(28). 1RA(8)). (1F1LE(29).
     31RA(9)). (IFILE(30). IRA(10)). (IFILE(31). IRA(11)). (IFILE(32).
     41RA(12)). (IFILE(33). IRA(13)). (IFILE(34). NBR). (IFILE(35). NC)
      EQUIVALENCE (XP(1.1), X(1)), (XXP(1.1), XX(1))
      DATA (ALPHA(1), I = 1, 5) (.20. .15. .10. .05. .01)
      DATA (TABLE(I \cdot I) + I = 1 \cdot 28)
     1(.300, .285, .265, .247, .233, .223, .215, .206, .199, .190, .183,
     2 .177. .173. .169. .166. .163. .160. .160. .160. .160. .160. .142.
     3 .142. .142. .142. .142. .131. .736)
      DATA (TABLE (1.2). 1 = 1, 28)
     1(.319. .299. .277. .258. .244. .233. .224. .217. .212. .202. .194.
     2 +187+ +182+ +177+ +173+ +169+ +166+ +166+ +166+ +166+ +147+
     3 -147, -147, -147, -147, -136, -768)
      DATA (TABLE (1.3). I = 1. 28)
     1(.352. .315. .294. .276. .261. .249. .239. .230. .223. .214. .207.
     2 •201 • •195 • •189 • •184 • •179 • •174 • •174 • •174 • •174 • •174 • •158 •
     3 .158. .158. .158. .158. .144. .805)
      DATA (TABLE (1.4), 1 = 1. 28;
     1(.381. .337. .319. .300. .285. .271. .258. .249. .242. .234. .227.
     2 .220, .213, .206, .200, .195, .190, .190, .190, .190, .190, .173,
     3 .173 .173 .173 .173 .173 .161 .886)
      DATA (TABLE (1.5). ! = 1. 28)
     1(.417, .405, .364, .348, .331, .311, .294, .284, .275, .268, .261,
     2 .257, .250, .245, .239, .235, .231, .231, .231, .231, .231, .200,
     3 .200, .200, .200, .200, .187, 1.031)
      CALL CRTID (4HKSA3)
      CALL GRS! (+06652+ +06652+ 1++ +06652)
      CALL SETIT
      CALL MOVE (0., XP, 5000)
      READ INPUT
  100 READ 110. JOB. FMT
  110 FORMAT (10A8/ 10A8)
      READ 120. IFILE
  120 FORMAT (12. 8x. 1311. 1x. 11. 5x. 411. 5x. 12. 2x. 1312. 4x. 12.
     12X • 11)
```

READ 130. JALPHA

```
130 FORMAT (6511)
    IF (NC) 140+ 160+ 140
140 READ 150. A. B. C. D. E
150 FORMAT (5F14.6)
160 K = 1
    KK = NBR
170 READ (MEDIUM, FMT) ITEST, (XX(1), 1 = K,KK)
180 IF (ITEST) 200: 190: 200
190 K = K + NBR
    KK = KK + NBR
    GO TO 170
200 N = K + ITEST -1
210 IRUN = 1
    TRANSFORMATION I
    IF (ITRAN(1)) 220, 240, 220
      DO 230 I = 1, N
220
230
       X(1) = XX(1)
      1T = 1
       ASSIGN 240 TO ILINE
       GO TO 740
    TRANSFORMATION 2
240 IF (ITRAN(2)) 250: 290: 250
250
      IT * 2
      ASSIGN 290 TO ILINE
      DO 280 I = 1 . N
      IF(XX(1)) 260, 280, 280
260
      PRINT 270. IT
      FORMAT (24H ERROR IN TRANSFORMATION, 3X, I2, 3X,21HLN OF NEGATIV
270
   1E NUMBER )
      GO TO 1330
280
      X(1) = ALOG(XX(1))
      GO TO 740
    TRANSFORMATION 3
290 IF (1TRAN(3)) 300, 330, 300
300
      ASSIGN 330 TO ILINE
      DO 320 I = 1 . N
      IF (XX(1)) 310. 320. 320
310
      PRINT 270. IT
      GO TO 1330
320
      x(1) = ALOG(ALOG(xx(1)))
      GO TO 740
    TRANSFORMATION 4
330 IF (ITRAN(4)) 340+ 370+ 340
340
      IT = 4
      ASSIGN 370 TO ILINE
      DO 360 I = 1 N
      ARG = A + XX(1)
      IF (ARG) 350. 360. 360
350
      PRINT 270. IT
      GO TO 1330
      X(1) = ALOG (ARG)
360
      GO TO 740
    TRANSFORMATION 5
370 IF (ITRAN(5)) 380, 420, 380
380
      IT = 5
      ASSIGN 420 TO ILINE
      DO 400 1 = 1. N
      ARG = C + \chi\chi(1)
```

```
IF (ARG) 410. 390. 390
  390
        ARG2 - B + ALOG(ARG)
        IF (ARG2) 410. 400. 400
  400
        X(I) = ALOG (ARG2)
        GO TO 740
        PRINT 270. IT
  410
        GO TO 1330
      TRANSFORMATION 6
  420 IF (ITRAN(6)) 430, 470, 430
        IT = 6
  430
        ASSIGN 470 TO ILINE
        DO 460 I = 1. N
        IF (XX(1)) 440+ 460+ 460
  440
        PRINT 450. IT
        FORMAT (24H ERROR IN TRANSFORMATION. 3X, 12. 3X. 32HSQUARE ROOT
  450
     10F A NEGATIVE NUMBER)
        GO TO 1330
  460
        X(I) = SQRT(XX(I))
        GO TO 740
C
      TRANSFORMATION 7
  470 IF (ITRAN(7)) 480: 520: 480
        IT = 7
  480
        ASSIGN 520 TO ILINE
        DO 510 I # 1, N
        IF (XX(1)) 510+ 490+ 510
        PRINT 500. IT
  490
  500
        FORMAT (24H ERROR IN TRANSFORMATION: 3X: 12: 3X: 11HZERO DIVIDE)
        GO TO 1330
  510
        X(I) = 1.07 XX(I)
        GO TO 740
      TRANSFORMATION B
  520 IF (ITRAN(8)) 530, 560, 530
  530
        1T = 8
        ASSIGN 560 TO ILINE
        DO 550 I = 1. N
        ARG = D + XX(1)
        IF (ARG) 550. 540. 550
  540
        PRINT 500. IT
        GO TO 1330
  550
        X(I) = 1.0/ARG
        GO TO 740
      TRANSFORMATION 9
  560 IF (ITRAN(9)) 570, 610, 570
  570
        IT = 9
        ASSIGN 610 TO ILINE
        DO 600 1 = 1. N
        ARG = 1. - XX(1) ## 2
        IF (ARG) 580, 580, 600
  580
        PRINT 590. IT
        FORMAT ( 24H ERROR IN TRANSFORMATION: 3X: 12: 3X: 47HZERO DIVIDE
     1 OR SQUARE ROOT OF A NEGATIVE NUMBER)
        GO '0 1330
  600
        X(1) = ATAN (XX(1) / SQRT (ARG))
        GO TO 740
      TRANSFORMATION 10
  610 IF (ITRAN(10)) 620, 650, 620
  620
        IT = 10
        ASSIGN 650 TO ILINE
        DO 640 1 = 1. N
```

```
ARG = 1. - XX(1)
        IF (ARG .GT. 0..AND. XX(1) .GE. 0.) GO TO 640
        PRINT 590. IT
  630
        GO TO 1330
        x(1) = 2. # ATAN (5QRT (XX(1))/SQRT(ARG))
 640
        GO TO 740
      TRANSFORMATION 11
C
  650 IF (ITRAN(11)) 660, 690, 660
        IT = 11
  660
        ASSIGN 690 TO ILINE
        DO 680 I = 1, N
        IF (E) 680, 670, 650
        PRINT 500. IT
  670
        GO TO 1330
  680
        X(1) = XX(1)/E
        GO TO 740
      TRANSFORMATION 12
  690 IF (ITRAN(12)) 700, 720, 700
  700
        DO 710 I = 1. N
        X(1) = SIN(XX(1))
  710
        IT = 12
        ASSIGN 720 TO ILINE
        GO TO 740
C
      TRANSFORMATION 13
  720
        DO 730 I = 1. N
        x(1) = COS(xx(1))
  730
         IT = 13
      PRINT INPUT
  740 PRINT 750. JOB. FMT
  750 FORMAT (1H1. 10A8/ 23H FORMAT FOR SAMPLE DATA: 5X. 10A8)
      PRINT 760, NRUN, IRUN, (IFILE(I), I= 2, 19)
  760 FORMAT (15H NO. OF RUNS = . 12. 5x. 7HIRUN = .
     121H TRANSFORMATION 1 = . II. 5x. 20HTRANSFORMATION
                                                             2 = .11.5x
     2 20HTRANSFORMATION 3 = , II. 5x. 20HTRANSFORMATION
                                                             4 = \cdot 11/
     321H TRANSFORMATION 5 = + 11. 5x. 20HTRANSFORMATION
                                                             6 = 11 \cdot 5X
                                                            8 = . 11/
     4 20HTRANSFORMATION 7 = . 11. 5x. 20HTRANSFORMATION
     521H TRANSFORMATION 9 = , II. 5x. 20HTRANSFORMATION 10 = , II. 5x.
     6 20HTRANSFORMATION 11 = . 11. 5x. 20HTRANSFORMATION 12 = . 11/
     721H TRANSFORMATION 13 = . 11. 5X. 9HMEDIUM = . 11/
     SISH PLOT TYPE A = . II.11x. 14HPLOT TYPE B = . II.11x. 14HPLOT TYP
     9E C = . 11.11X. 14HPLOT TYPE D = . [1]
      PRINT 770, (IFILE(1), 1 = 20, 35), ITEST, IALPHA
  770 FORMAT ( 9H NOTIC . 12. 16x. 6HIRA . 13( 12. 1X)/ 7H NBR .
     112.18x. SHNC . . 11. 20x. BHITEST . . 12/24H ALPHA IDENTIFICATION
     2= . 13(5[1. 1X))
      IF (NC) 780, 800, 780
  780 PRINT 790. A. B. C. D. E
  790 FORMAT (29H TRANSFORMATION CONSTANT CARD: 3X: 4HA = + F14.6: 3X:
     1 4H8 = \cdot F14.6. 3X. 4HC = \cdot F14.6/32X. 4H0 = \cdot F14.6. 3X. 4HE = \cdot
     2 F14.61
  800 NM = N/2 + MOD (N+2)
      NN = 0
  210 NK # NN + 1
      NN = MINO (50: NM) +NN
      PRINT ORIGINAL AND TRANSFORMED DATA
C
      PRINT 820. ((XXP(J.1), XP(J.1), J = 1.2). [ =NK.NN)
  820 FORMAT (1H1. 9x. 13HORIGINAL DATA: 15x. 16HTRANSFORMED DATA: 16x.
      113HORIGINAL DATA, 15x. 16HTRANSFORMED DATA/ (5x. F20.10. 10x. F20.
      210. 10x. F20.10.
                        10x, F20.10))
```

```
NM = NM - 50
      if (NM) 830, 830, 810
      SORT DATA
 830 K = N - 1
     D9 850 I = 1 . K
      11 = 1 + 1
      DO 850 J = 11. N
      IF (X(1) - X(J)) 850, 850, 840
  840 TEMP = X(J)
      X(J) = X(I)
      X(1) * TEMP
  850 CONTINUE
      ESTABLISH PARAMETERS FOR PLOTS
C
      XMI = X(I)
      XMA = X(N)
      RANGE = XMA - XMI
      IF (NUTIC .EQ. 0) NOTIC = 15
      DX = RANGE / FLOAT (NOTIC)
      TEN (1) = 10.E-10
      IF (DX .GT. TEN(1)) GO TO 860
      DX = TEN(1)
      GO TO 900
  860 IF (DX +LT+ 10+E10) GO TO 870
      DX = 10.E10
      GO TO 900
  870 DO 880 M = 2. 20
      TEN (M) = 10.44 (M - 11)
      IF (DX +GT+ TEN(M-1) .AND.DX .LE. TEN(M)) GO TO 890
  880 CONTINUE
  890 NDX = DX/ TEN(M-1) + .9
      DX = FLOAT (NDX) + TEN(M+1)
  900 DELTA = RANGE/ 500.
      FLN = FLOAT(N)
      SUM = 0.
      SUMSQ = 0.
C
      COMPUTE AND PRINT MEAN ADD VARIANCE
      DO 910 I = 1. N
      SUM = SUM + X(I)
  910 SUMSQ = SUMSQ + X(I) + X(I)
      XBAR = SUM/ FLN
      SIGMA2 = (SUMSQ - FLN + XBAR ++ 2)/ (FLN - 1.)
      SIGNA = SQRT (SIGMA2)
      PRINT 920, XMI, XMA, RANGE, N. XBAR, SIGMA2, SIGMA
  920 FORMAT (1H0. BX. 10HMINIMUM = . E15.8. 20%. 10HMAXIMUM = . E15.8.
     110X. BHRANGE = . E15.8/ 5X. 14HSAMPLE SIZE = . 15. 33X. 7HMEAN =
     2. E15.8/ 8x. 11HVARIANCE = . E15.8. 9x. 21HSTANDARD DEVIATION = .
     3E15.8)
      COMPUTE EMPIRICAL COF
      DD = 0.
       J = 1
      IFREG (J) = 0
      DO 990 I = 1. N
      IFREQ(J) = IFREQ(J) + 1
      IF (X(1) - X(1+1)) 930+ 990+ 930
  930 \times (J) = \times (I)
      U(J) = (X(J)
                   - XBAR // SIGMA
      F(J) = FREQ(U(J))
      S(J) = FLOAT(I)/ FLOAT (N)
С
      DETERMINE TEST STATISTIC D
```

```
SN(J) = ABS (F(J) - 5(J))
      IF (J - 1) 950, 940, 950
  940 SN1 (1) = ABS (F(J)
      GO TO 960
  950 SN1 (J) = ABS (F(J)
                           S(J -1))
  960 ZMAX = AMAX1 (SN(J)
                          SNI(J))
      IF (ZMAX - DD) 980
                           180. 970
  970 JJ = J
      DD = ZMAX
  980 J = J + 1
      IFREQ (J) = 0
  990 CONTINUE
      PRINT STATISTICS
      J = J - L
      11 = J
      IK . 0
 1000 IL = IK + 1
      1K = MINO (50 + 11) + 1K
      PRINT 1010
 1010 FORMAT (1H1, 54x, 29HKOLMOGOROV TEST FOR NORMALITY//3x,
                                                                      11HOR
     IDER INDEX. 6x. 11HTRANSFORMED.18x. 12HSTANDARDIZED. 1x.
     2 15HTHEORETICAL CDF. 2x. 12HOBSERVED CDF. 3x.
                                                                14HABSOLUT
     3E VALUE: 3X: 14HABSOLUTE VALUE/ 7X: 3H(J): 11X: 9HDATA X(J):
     4 7x. 9HFREQUENCY. 4x.
     512HVARIATE U(J) . 5X. 4HF(J) . 12X. 4HS(J) . 9X. 11HF(J)-S(J-1), 7X.
     69HF (J)-5(J)/)
      PRINT 1020 ((1)
                        X(1). IFREQ (1). U(1). F(1). S(1). SN1(1). SN(1)
     1) . I = IL . IK)
 1020 FORMAT
                   (6X: I4: 6X: F14:6: 10X: I4: 4X: F14:11: 1X: F14:12:
     1 2x. F14.12. 1x. F14.13. 2x. F14.13)
      11 = 11 - 50
      IF (II) 1030, 1030, 1000
 1030 PRINT 1040. JJ. DD
 1040 FORMAT ( 3H0D(*, 14, 4H) = *, F7.4)
      T(1) = XMI
C
      COMPUTE THEORETICAL CDF
      DO 1050 L = 1. 500
      UV = (T(L) - XBAR)/SIGMA
      FN(L) = FREQ(UV)
      IF (T(L) .GT. XMA) GO TO 1060
      T(L+1) = T(L) + DELTA
 1050 CONTINUE
 1060 DO 1320 II = 1. 5
      OBTAIN VALUE FOR D SUB ALPHA OF N
      IF (IALPHA (II. IT)) 1070, 1320, 1070
 1070 IF (IA .EQ. 0 .AND. IB .EQ. 0 .AND. IC .EQ. 0) GO TO 1150
      IF (N - 30) 1080, 1080, 1090
 1080 NN R N - 3
      DALPHA = (TABLE (NN. II))
      GO TO 1100
 1090 DALPHA = TABLE (28. II)/ SQRT(N)
 1100 IF (IB .EQ. O .AND. IC .EQ. 0) GO TO 1120
      COMPUTE CONFIDENCE LIMITS
      DO 1110 I = 1, J
      SU(1) = S(1) + DALPHA
 1110 SL(1) = S(1) - DALPHA
 1120 IF (IA) 1130, 1150, 1130
      COMPUTE REJECTION REGION
C
 1130 DO 1140 I = 1, 500
```

```
FU(1) =FN(1) + DALPHA
1140 FL(1) =FN(1) - DALPHA
C*4 ***P L O T S
     HEADINGS
1150 00 1310 KK # 1+4
      IF (IFILE (KK + 15)) 1160. 1310. 1160
1160 PRINT 1170. JOB. IRUN. N
1170 FORMAT (2M$2. 15X. 10A8/ 1M$. 15X. 11HRUN NUMBER . 12. 6H WITH .
     115. 13H OBSERVATIONS)
      GO TO (1180+ 1200+ 1220+ 1240)+ KK
      HEADINGS FOR PLOT TYPE A
1180 PRINT 1190: ALPHA(II): DALPHA
 1190 FORMAT (1H$415X4 39HOBSERVED RELATIVE CUMULATIVE FREQUENCY+/ 1H8+
     115%, 46HTHEORETICAL RELATIVE CUMULATIVE FREQUENCY: AND/ 1Hs.
     215X. 34HCRITICAL REGION BOUNDS FOR ALPHA =. F5.2. 3X.
     22HLEVEL OF SIGNIFICANCE./1Hs. 15x. 11HD(ALPHA) = . F5.3)
      GO TO 1260
      MEADINGS FOR PLOT TYPE B
 1200 JCB = 100. + (1. - ALPHA(II)) + .5
      PRINT 1210. ICB. DALPHA
 1210 FORMAT (1H$.15x. 39HOBSERVED RELATIVE CUMULATIVE FREQUENCY./ 1H$.
     115X, 42HTHEORETICAL RELATIVE CUMULATIVE FREQUENCY./
                                                                1H$ .
     2:15X. 12.1X.78HPERCENT CONFIDENCE BANDS FOR THE THEORETICAL CUMULAT
     GIVE DISTRIBUTION FUNCTION:/ 148:15X: 11HD(ALPHA) = : F5:3)
      GO TO 1260
      MEADINGS FOR PLOT TYPE C
 1220 ICB = 100. + (1. - ALPHA(11)) + .5
      PRINT 1230. 1CB
 1230 FORMAT (1H$.15%, 46HOBSERVED RELATIVE CUMULATIVE FREQUENCY AND THE
     1/ 1MS.15x.12. 1x.77HPERCENT CONFIDENCE BANDS FOR THE THEORETICAL C
     2UMULATIVE DISTRIBUTION FUNCTION)
      GO TO 1260
      HEADINGS FOR PLOT TYPE D
 1240 PRINT 1250
 1250 FORMAT ( 1HS.15X. 54HOBSERVED AND THEORETICAL RELATIVE CUMULATIVE
     IFREQUENCY)
PLOT MEAN AND S(X)
 1260 CALL GRF (XMI: 0:: XMA: 1:: XMI: 0::DX::05:IRA(IT):2:0: 1: XBAR:
     1 .5. 1. 1. 2H12. -1. 0)
      CALL GRF (XMI, 0., XMA, 1., XMI, 0., DX, 05, IRA(IT), 2.1, J. X(1),
     1 5(1), 1, 1, 2H75, 0, 0)
      GO TO (1270+ 1280+ 1290+ 1300)+ KK
      ADDITIONAL CURVES FOR PLOT TYPE A
 1270 CALL GRF (XMI: 0: XMA: 1: XMI: 0: DX::05:[RA(IT):2:1:1000:T(1):
     1FN(1) + 1 + 1 + 2H75 +-1 + 0)
      CALL GRF (XMI+ 0++ XMA+ I++ XMI+ 0++ DX++05+1RA(IT)+2+1+1000+T(1)+
     1FL(1) + 1 + 1 + 2H75 +-1 + 0)
      CALL GRF (XMI: 0: XMA: 1: XMI: 0: DX: 05: IRA(IT):2:1:1000:T(1):
     IFU(1) . 1 . 1 . 2H75 -1 . 0)
      GO TO 1310
      ADDITIONAL CURVES FOR PLOT TYPE B
 1280 CALL GRF (XM): 0.. XMA: :.. XMI: 0.. DX:.05: IRA(IT):2:1:1000:T(1):
     1FN(1) . 1 . 1 . 2H75 -- 1 . 0)
      CALL GRF (XMI: 0:: >60: 1:: XMI: 0:: DX::05:IRA(IY):2:1: J: X(1):
     ISL(1) . 1 . 1 . 2H75 . 0 . 0)
      CALL GRF (XMI, 0., XMA, 1., XMI, 0., DX, 05, IRA(IT), 2.1. J, X(1),
     ISU(1) + 1 + 1 + 2H75 + 0 + 0)
      GO TO 1310
C
      ADDITIONAL CURVES FOR PLOT TYPE C
```

```
1290 CALL GRF (XMI. 0.. XMA. 1.. XMI. 0.. DX..05.1RA(IT).2.1. J. X(I).
     ISL(1) . 1 . 1 . 2H75 . 0 . 0)
      CALL GRF (XMI, 0.. XMA: 1.. XMI: 0.. DX: 05: IRA(IT): 2:1: J: X(1):
     ISU(1), 1, 1, 2H75, 0, 0)
      GO TO 1310
      ADDITIONAL CURVES FOR PLOT TYPE D
 1300 CALL GRF (XMI+ 0++ XMA+ 1++ XMI+ 0++ DX++05+19A(17)+2+1+1000+T(1)+
     1FN(1). 1. 1. 2H75.-1. 0)
1310 CONTINUE
1320 CONTINUE
1330 IRUN = IRUN + 1
      IF (IRUN - NRUN) 1340, 1340, 1350
1340 GO TO ILINE
1350 CALL INTVL (Z)
1360 FORMAT (19HOTIME IN SECONDS = . F10.2)
      PRINT 1360. Z
      GO TO 100
      END
         SUBTYPE . DATA
T
```

## VIII. REFERENCES

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Following a brief discussion on the background, applicability, and limitations of the Kolmogorov Test, a computer program of the Kolmogorov Test for normality is described. The program is applicable for testing the null hypothesis that a random sample is from a parent normal population whose mean and variance are equal to those of the sample distribution. The program determines the minimum and maximum sample values, computes sample estimates of the mean and variance of the hypothesized normal distribution, and computes the Kolmogorov statistic. Five optional significance levels (0.20, 0.15, 0.10, 0.05, and 0.01) are available, and sample size limitations are 4 5 n \$2500. An optional feature provides for CRT plot output applicable for both hypothesis testing and interval estimation.

The program is coded in FORTRAN IV for the IBM 7050 (SIREICH) computer,

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